

냉각 결정화 기술의 응용전략-2

고려대 화공생명공학과 양대륙

결정화공정의 모델링

결정화 모델의 구성 및 활용

Model for batch cooling crystallization

o Model for the crystallizer ¾For modeling crystallization process, PBE is employed.

$$
\frac{\partial f(L,t)}{\partial t} = -\frac{\partial [G(L,t) \cdot f(L,t)]}{\partial L} \qquad \qquad f(0,t) = -\frac{B^0}{G\Big|_{L=0}}
$$

where, *f* is the population density of crystals, *^t* is the time, *L* is the size of the particles, and G is the growth rate. (No breakage or agglomeration)

 \blacktriangleright The overall growth rate is given by

$$
G = k_g(T, L) \cdot \Delta c^g
$$

where Δ*c* is the difference of the concentration in the solution, *^c*, and the saturation concentration c^* . And $K_{\!\scriptscriptstyle g}$ is the growth kinetic parameter.

Model for batch cooling crystallization

\triangleright The growth kinetic parameter, k_g

$$
k_g(T, L) = k_0 \exp(\frac{-E_g}{RT})(1 + k_1 L)^{k_2}
$$

where, *f* is the population density of crystals, *^t* is the time, *L* is the size of the particles,and G is the growth rate.

\triangleright The mass balance

$$
\frac{dc}{dt} = -\rho \cdot k_v \frac{d\mu_3}{dt}
$$

where, $\,\mu_{\,\mathbf{3}}$ is the third moment of the CSD

▶ The third moment of the CSD
$$
\frac{8}{x}
$$

$$
\mu_3 = \int_0^\infty n \cdot L^3 dL
$$

¾

Model for batch cooling crystallization

 \triangleright The nucleation rate

$$
B^0 = k_b(T) \cdot \Delta c^{\beta}(T)
$$

The nucleation occurs if the metastable limit is violated.

 \triangleright The population density of crystals is expressed as following equation,

The method which was suggested by Hu et al. [2] provides ^a means to reduce the PBE to a set of algebraic equations.

$$
f(L_i, t_{j+1}) \approx \frac{f(L_i, t_j)}{1 + \frac{\partial G(L, t)}{\partial L}\Big|_{L=L_i} \Delta t}
$$

 \blacktriangleright $L_{i,j+1}$ is given by

$$
L_{i,j+1} \approx G(L_{i,j},t_j)\Delta t + L_{i,j}
$$

Simulation results of static MSL model with PBE

Crystal size distribution of static MSL model (CV = 34%)

Dynamic MSL model parameters

1.The parameter of MSL model are function of saturation temperature.

2.The value of parameter 'p' is fixed at 2.2 from experimental results.

300.0012498 0.0316436

Relation of model parameter & saturation temperature

대해 전화
[편태]

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Simulation results of dynamic MSL model

Simulation results of dynamic MSL model

| 대유|| 정의
|
| 역리|

Crystal size distribution of dynamic MSL model (CV = 16%)

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Crystallization Applet

Crystallization applet

- \triangleright Developed for enhancing the understandings of the crystallization behavior.
	- **Effect of kinetic parameters**
	- **Effect of agglomeration and breakage**
	- Effect of cooling strategy
- \triangleright Applet program
	- Platform independent
	- Java runtime is required
	- **Interactive user interface**
	- No user installation
	- Online/offline

Crystallization Applet

CRYSTALLIZATION APPLET I

최적 냉각곡선의 도출

냉각결정화의 냉각 전략

기존 이론

Process System Engineering Lab., Dept. of Chemical and Biological Engineering, Korea University l T

- * Controlled cooling and seeding increases the weight mean size.
- * The width of the distribution increases

Optimal Cooling Strategy

o Inside metastable region

 $\Delta T_{\rm max} > \Delta T > 0$

As possible as close to the metastable limit

 \triangleright The higher supersaturation makes the higher crystal growth rate.

Desired supersaturation level ($\eta_\textrm{\tiny{des}}$)

- \triangleright The nucleation should be prohibited.
- \triangleright Too close to the limit is dangerous.
- ¾ \blacktriangleright Desired supersaturation level $\eta_\textrm{\tiny{des}}$ <1.

$$
\eta = \frac{\Delta T}{\Delta T_{\text{max}}}
$$

Objective function for optimal cooling curve

- ¾Multi-step prediction is used.
- \triangleright Objective is maximizing mean crystal size.
- \blacktriangleright Violation over desired supersaturation level η_{des} works as penalty.

$$
\min_{T(t_k)} -\omega_1 L(t_N) + (\eta - \eta_{des})^T W(\eta - \eta_{des})
$$

if $\eta_k - \eta_{des} < 0$, then $\eta_k - \eta_{des} = 0$
where $\eta = \frac{T^*(C) - T}{\Delta T_{max}} = \frac{\Delta T}{\Delta T_{max}}$

Optimization based on Genetic Algorithm

Characteristics of genetic algorithm (GA)

- \blacktriangleright GA is a search technique used in computing to find exact or approximate solutions to optimization and search problems.
- \triangleright Genetic algorithms are categorized as global search heuristics.
- \triangleright Genetic algorithms are a particular class of evolutionary algorithms that use techniques inspired by evolutionary biology such as inheritance, mutation, selection, and crossover.

Advantages of GA

- \triangleright The major advantage of GA is their flexibility and robustness as a global search method.
- \triangleright They do not need gradient information and make relatively few assumptions about the problem being solved.
- \triangleright They can deal with highly nonlinear problems and nondifferentiable functions as well as functions with multiple local optima.

Optimization based on genetic algorithm

a Fitness function

- \triangleright In order that the crystal growth rate is maximized while the nucleation rate is minimized, the operation should be close to MSL as possible and this problem can be formulated as an optimization problem.
- \triangleright The objective function can be chosen as a function of the third moments of the CSD.
- \triangleright For reality, cooling rate has upper and lower bound.
- \triangleright Operation time and termination temperature sets to be identical for all types of the cooling strategies.

$$
\min_{u(t)} \quad (\mu_3^n + w \frac{1}{\mu_3^n})
$$
\n
$$
\text{subject to} \quad 0 \le u(t) \le 50
$$

Simulation Results (Linear cooling)

Fig 2. (a) Evolution of CSD for the linear cooling curve, (b) The linear cooling curve and metastable limit, (c) Seed and newly formed crystal size distribution

Simulation Results (Natural cooling)

Fig 3. (a) Evolution of CSD for the natural cooling curve, (b) The natural cooling curve and metastable limit, (c) Seed and newly formed crystal size distribution

Simulation Results (Optimal cooling)

Fig 1. (a) Evolution of CSD for the optimal cooling curve, (b) The optimal cooling curve and metastable limit, (c) Seed and newly formed crystal size distribution

Results

Fig 3. Comparison of three types of the cooling curve (optimal, linear, natural curve)

Example 1: Linear cooling

Experimental procedure \blacksquare

- (1) Making up solution.
	- (NH4)2SO4 **–** H2O solution
	- Concentration : 0.8425(Ts=50℃)
- ② Keeping temperature as initial temperature for 1hr.
	- RPM of agitator : 1000rpm
- ③ Starting cooling experiment
- ④ Adding the seed crystal when reactor temperature cross over the saturation temperature
	- Seed crystal size: 462.5µm
	- Seed crystal weight : 10g
- ⑤ Filtering solution and drying the crystal

Estimating optimal initial temperature for linear cooling

- \triangleright Experimental condition
	- Solution concentration : 0.8425 [solute kg/ solvent kg] (T_s=50℃)
	- Cooling rate : 20℃/h
	- **Processing time : 80 min.**
	- initial temperature : 51℃, 53℃, 55℃, 57℃
- \triangleright Expected results
	- From 51℃
		- \checkmark Crystal size : 1050 $\rm \mu m$
		- \checkmark Total weight : 92.61g
		- \checkmark Broad crystal size distribution.
	- From 55℃
		- \checkmark Crystal size : 990 μ m
		- \checkmark Total weight : 77.62g
		- \checkmark Narrow crystal size distribution.

Simulation results of linear cooling from 51℃ and 53℃

Simulation results of linear cooling from 55℃ and 57℃ \bullet

nnh n

Crystal size distribution Accumulated weight percent

1400 1600 1800 2000

1200 1000

Crystal size [pm]

[⎢] [⎥] ⎢⎣ ⎥⎦ ※ CV (Coefficient of variance) under the 20% Implies uniform size distribution.

$$
CV = 100 \left[\frac{L_{w_{84\%}} - L_{w_{16\%}}}{2L_{w_{50\%}}} \right]
$$

Example 2: Linear vs. Optimal

o Optimal cooling Strategy

- ¾ Experimental condition
	- Solution concentration : 0.8425 [solute kg/ solvent kg]
	- **Processing time : 80 min.**
	- Initial temperature : 52.5℃
	- Final temperature : 23℃
	- Cooling curve
		- 9 Linear cooling : 22.125℃/h
		- \checkmark Optimal cooling : by simulated data
			- » Maximum cooling rate : 25℃/h
			- » Maximum heating rate : 5℃/h

\triangleright Expected results

- Optimal cooling
	- \checkmark Crystal size : 1050 $\rm \mu m$
	- \checkmark Total weight : 92.61g
	- \checkmark Narrow crystal size distribution.
- **Linear cooling**
	- \checkmark Crystal size : 1050 $\rm \mu m$
	- \checkmark Total weight : 92.61g
	- \checkmark Broad crystal size distribution.

(a) Optimal cooling line, (b) Linear cooling line

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Process Analytical Technology (PAT)의 응용

PAT

Process Analytical Technology (PAT)란?

- ¾ 온라인 측정기를 이용한 결정화 운전 및 제어기술
- ¾ PAT의 목적은 생산 공정을 이해하고 제어하기 위함.
- 왜 중요한가?
	- ¾ **"**Quality can not be tested into products; it has to be built in by design.**" –** FDA

제품이 동일한 과정에 의해 생산되고 품질이 보증 되어야 함.

- PAT의 기대효과
	- ¾ 효율향상 ¾ 원가절감 ¾ 균일한 품질

PAT Tool의 기대효과

\triangleright Multivariate data acquisition/analysis

중요 공정변수를 파악하기 위한 Design of Experiment

▶ Process analyzers

Off-line, at-line, on-line, in-line, non-invasive

\triangleright Process endpoint monitoring and control

공정을 모니터링하고 원하는 조업조건에 유지

최종 생산물에 대해 반응 또는 결정화 시간 대신 보다 명확한 물성을 이용한 결정 가능

¾ Continuous improvement/knowledge management

사후 관리 및 공정개선의 과학적 자료 축적

e Process Analyzers

- ¾ 전통적인 측정 방법
	- Temperature, pressure, pH, turbidity, … probes
	- **Mass flow meter**
	- Volumetric gas uptake/evolution

▶ 최근의 측정방법

- In situ real-time (Operando) spectroscopy (midIR, NIR, UV-vis, Raman, acoustic)
- *In situ* real-time particle analyzer (FBRM, PVM,...)

¾미래에 나올 방법들

॥ ⊥...

- **Advanced data management** software package
- **Remotely controlled** in situ real-time process sensors with high sensitivity
- **New process sensors (combo probe,** diode laser frequency modulation spectroscopy, …)

결정의 분포 측정

¾ Conventional methods (Offline)

- Sieving: 5–12500 μ m
- Microscopy: 0.5 –150 μ m

¾ In-process Ultrasonic spectroscopy

주파수가 1-150MHz 범위의 초음파가 1mm-5cm의 path를 통과하는 동안에 고체의 농도에 따라 초음파의 강도(intensity) 가 변화하는 것을 측정

결정 크기가 0.1μm-100μm 사이인 경우 30%-70% 사이의 solid 농도 측정

온도별, 농도별 강도의 Attenuation에 따른 calibration 필요

► In Situ Particle Size Analyzer (PSA)

- Laser beam을 조사하여 반사되는 것을 측정해 입자의 수나 크 기를 측정 (Focused Beam Reflectance Measurement, FBRM)
- 입자의 형태에 따라 어려운 calibration 필요 절대적 측정값보다는 상대적 변화의 측정에 적합

¾ In-process Video Microscopy (PVM)

- 반응기 내의 한 지점에서 시간에 따른 결정 사진 capture
- 결정의 크기나 모양의 변화, agglomeration, breakage등을 눈으로 확인
- 자체적인 측정값을 내보내지 않음
- *ln situ* PSA와 같이 사용하는 것이 일반적
- 예) 10분 간격으로 측정된 그림

결정구조 및 Morphology의 측정

- **▶ Raman spectroscopy**
	- 결정 구조 측정
- ¾ In-process XRD (X-Ray Diffraction)
	- Polymorph에 따라 다른 위치에서 나타나는 X-ray 회절의 강 도에 의해 polymorph의 농도 측정
- $>$ PVM
	- 반응기내의 결정에 대한 사진을 통해 결정

Wavenumber

 35° C

대표적 Example

유효성분 (API)이 두가지 polymorphs 를 가짐

- ▶ Drowning-out에 의해 생산
- ▶ Form I0| Form II보다 안정적

공정상 문제점

- ¾ 불가피한 Form II의 생성 (전형적으로 2~10%)
- ¾ 매우 느린 Form I 성장속도 (기존 방법으로 18 시간 이상 소요)
- ¾ 매우 느린 Form II에서 Form I로의 변환 속도 (특정용액에서 몇일 정도)
- ▶ 40ºC이상에서 분해 (낮은 온도에서 조업)
- ¾ Form I 세척에 많은 세척수 필요 (<500 L/m2/hr)

공정개선의 주안점

- ¾ Form I의 선택적 성장을 위한 중요 파라미터의 파악
- ¾ 순수한 Form I의 생산을 위한 강건하고 효율적인 공정의 설계

Form I seed + controlled n-Heptane addition at 25°C

Control anti-solvent Addition Rate to maintain $C < C_n$

Narrow metastable zone of Form II

- ¾ 중요 파라미터 도출
	- Anti-solvent addition rates
	- **Solvent composition**
	- **Seed quality and loading**

개선결과

- <mark>▷ 확대된 순수한 Form I 생산 영역</mark>
- ¾ Cycle time 단축 (2~4 hours vs. 18 hours)
- ¾ 여과효율 향상 (~ 400 L/m 2/hr vs. <500 L/m 2/hr)
- ¾ 단위부피당 생산성 향상 (~60 g/L vs. ~20 g/L)

Thank you!

